

# ON A SURPRISING RELATION BETWEEN THE MARCHENKO-PASTUR LAW, RECTANGULAR AND SQUARE FREE CONVOLUTIONS

FLORENT BENAYCH-GEORGES

**ABSTRACT.** In this paper, we prove a result linking the square and the rectangular  $R$ -transforms, the consequence of which is a surprising relation between the square and rectangular versions of the free additive convolutions, involving the Marchenko-Pastur law. Consequences on random matrices, on infinite divisibility and on the arithmetics of the square versions of the free additive and multiplicative convolutions are given.

Dans cet article, on prouve un résultat reliant les versions carré et rectangulaire de la  $R$ -transformée, qui a pour conséquence une relation surprenante entre les versions carré et rectangulaire de la convolution libre additive, impliquant la loi de Marchenko-Pastur. On donne des conséquences de ce résultat portant sur les matrices aléatoires, sur l'infinie divisibilité et sur l'arithmétique des versions carré des convolutions additives et multiplicatives.

## CONTENTS

Introduction	2
1. A relation between the Marchenko-Pastur law, the square and the rectangular free convolutions	3
1.1. Prerequisites on square and rectangular analytic transforms of probability measures	3
1.2. A relation between the square and the rectangular $R$ -transforms	4
1.3. Main result of the paper	6
2. Consequences on square and rectangular infinite divisibility	7
2.1. Prerequisites on infinite divisibility and Lévy-Kinchine formulas	7
2.2. Main result of the section	8
References	10

---

*Date:* July 9, 2009.

*2000 Mathematics Subject Classification.* 46L54, 15A52.

*Key words and phrases.* Free Probability, Random Matrices, Free Convolution, Infinitely Divisible Laws, Marchenko-Pastur Law.

This work was partially supported by the *Agence Nationale de la Recherche* grant ANR-08-BLAN-0311-03.

## INTRODUCTION

Free convolutions are operations on probability measures on the real line which allow to compute the empirical spectral<sup>1</sup> or singular<sup>2</sup> measures of large random matrices which are expressed as sums or products of independent random matrices, the spectral measures of which are known. More specifically, the operations  $\boxplus, \boxtimes$ , called respectively *free additive and multiplicative convolutions* are defined in the following way [VDN91]. Let, for each  $n$ ,  $M_n, N_n$  be  $n$  by  $n$  independent random hermitian matrices, one of them having a distribution which is invariant under the action of the unitary group by conjugation, the empirical spectral measures of which converge, as  $n$  tends to infinity, to non random probability measures denoted respectively by  $\tau_1, \tau_2$ . Then  $\tau_1 \boxplus \tau_2$  is the limit of the empirical spectral law of  $M_n + N_n$  and, in the case where the matrices are positive,  $\tau_1 \boxtimes \tau_2$  is the limit of the empirical spectral law of  $M_n N_n$ . In the same way, for any  $\lambda \in [0, 1]$ , the *rectangular free convolution*  $\boxplus_\lambda$  is defined, in [B09], in the following way. Let  $M_{n,p}, N_{n,p}$  be  $n$  by  $p$  independent random matrices, one of them having a distribution which is invariant by multiplication by any unitary matrix on any side, the symmetrized<sup>3</sup> empirical singular measures of which tend, as  $n, p$  tend to infinity in such a way that  $n/p$  tends to  $\lambda$ , to non random probability measures  $\nu_1, \nu_2$ . Then the symmetrized empirical singular law of  $M_{n,p} + N_{n,p}$  tends to  $\nu_1 \boxplus_\lambda \nu_2$ . These operations can also, equivalently, be defined in reference to free elements of a non commutative probability space, but in this paper, we have chosen to use the random matrix point of view.

In the cases  $\lambda = 0$  or  $\lambda = 1$ , i.e. where the rectangular random matrices considered in the previous definition are either “almost flat” or “almost square”, the rectangular free convolution with ratio  $\lambda$  can be expressed with the additive free convolution:  $\boxplus_1 = \boxplus$  and for all symmetric laws  $\nu_1, \nu_2$ ,  $\nu_1 \boxplus_0 \nu_2$  is the symmetric law the push-forward by the map  $t \mapsto t^2$  of which is the free additive convolution of the push forwards of  $\nu_1$  and  $\nu_2$  by the same map. These surprising relations have no simple explanations, but they allow to hope a general relation between the operations  $\boxplus_\lambda$  and  $\boxplus$ , which would be true for any  $\lambda$ . Up to now, despite many efforts, no such relation had been found, until a paper of Debbah and Ryan [DR07], where a relation between  $\boxplus_\lambda, \boxplus$  and  $\boxtimes$  is proved in a particular case. In the present paper, we give a shorter proof of a wide generalization<sup>4</sup> of their result: for any  $\lambda \in (0, 1]$ , we define  $\mu_\lambda$  to be the law of  $\lambda$  times a random variable with law the Marchenko-Pastur law with mean  $1/\lambda$ , and we prove that for any pair  $\mu, \mu'$  of probability measures on  $[0, +\infty)$ , we have

$$(1) \quad \sqrt{\mu \boxtimes \mu_\lambda} \boxplus_\lambda \sqrt{\mu' \boxtimes \mu_\lambda} = \sqrt{(\mu \boxplus \mu') \boxtimes \mu_\lambda},$$

where for any probability measure  $\rho$  on  $[0, +\infty)$ ,  $\sqrt{\rho}$  denotes the symmetrization of the push-forward of  $\rho$  by the map  $t \mapsto \sqrt{t}$ . Our proof is based on the following relation between the  $R$ -transform<sup>5</sup>  $R_\mu$  of a probability measure  $\mu$  on  $[0, +\infty)$  and the rectangular  $R$ -transform  $C_{\sqrt{\mu \boxtimes \mu_\lambda}}$

<sup>1</sup>The *empirical spectral measure* of a matrix is the uniform law on its eigenvalues with multiplicity.

<sup>2</sup>The *empirical singular measure* of a matrix  $M$  with size  $n$  by  $p$  ( $n \leq p$ ) is the empirical spectral measure of  $|M| := \sqrt{MM^*}$ .

<sup>3</sup>The *symmetrization* of a law  $\mu$  on  $[0, +\infty)$  is the law  $\nu$  defined by  $\nu(A) = \frac{\mu(A) + \mu(-A)}{2}$  for all Borel set  $A$ . Dealing with laws on  $[0, +\infty)$  or with their symmetrizations is equivalent, but for historical reasons, the rectangular free convolutions have been defined with symmetric laws. In all this paper, we shall often pass from symmetric laws to laws on  $[0, +\infty)$  and vice-versa. Thus in order to avoid confusion, we shall mainly use the letter  $\mu$  for laws on  $[0, \infty)$  and  $\nu$  for symmetric ones.

<sup>4</sup>See Remark 5.

<sup>5</sup>Note that there are two conventions regarding the  $R$ -transform. The one we use is the one used in the combinatorial approach to freeness [NS06], which is not exactly the one used in the analytic approach [HP00]:  $R_\mu^{\text{combinatorics}}(z) = z R_\mu^{\text{analysis}}(z)$ .

with ratio  $\lambda$  of  $\sqrt{\mu \boxtimes \mu_\lambda}$ : we prove that for all  $z$ ,

$$R_\mu(z) = C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z).$$

This relation also allows us to prove precise relations between  $\boxplus$ -infinitely divisible laws and  $\boxplus_\lambda$ -infinitely divisible laws.

We would like to observe that formula (1) has some consequences which are far from obvious. It means that for  $n, p$  large integers such that  $n/p \simeq \lambda$ , for  $A, B, M, M'$  independent random matrices with respective sizes  $n \times n$ ,  $n \times n$ ,  $n \times p$  and  $n \times p$  such that  $A, B$  are invariant in law under left and right multiplication by unitary matrices and  $M, M'$  have independent Gaussian entries, then as far as the spectral measure is concerned,

$$(AM + BM')(AM + BM')^* \simeq AM(AM)^* + BM'(BM')^*.$$

It also means, if  $1 \ll n \ll p$ , that for  $X, Y$  independent  $n \times p$  random matrices, as far as the spectrums are concerned,

$$(X + Y)(X + Y)^* \simeq XX^* + YY^*.$$

The relation (1) has also consequences on the arithmetics of free additive and multiplicative convolutions  $\boxplus$  and  $\boxtimes$  (Corollaries 7 and 12) which wasn't known yet, despite the many papers written the last years about questions related to this subject, e.g. [BV95, BPB99, B04, CG08a, CG08b, BBG08, BBCC08].

**Acknowledgments:** The author would like to thank his friend Raj Rao for bringing the paper [DR07] to his attention and Øyvind Ryan and Serban Belinschi for some useful discussions. He would also like to thank an anonymous referee for suggesting him Remark 8.

## 1. A RELATION BETWEEN THE MARCHENKO-PASTUR LAW, THE SQUARE AND THE RECTANGULAR FREE CONVOLUTIONS

### 1.1. Prerequisites on square and rectangular analytic transforms of probability measures.

1.1.1. *The square case: the  $R$ - and  $S$ -transforms.* These are analytic transforms of probability measures which allow to compute the operations  $\boxplus$  and  $\boxtimes$ , like the Fourier transform for the classical convolution. The  $R$ -transform can be defined for any probability measure on the real line, but we shall only define it for probability measures on  $[0, +\infty)$ . Consider such a probability measure  $\mu$ . If  $\mu = \delta_0$ , then  $R_\mu = S_\mu = 0$ . Now, let us suppose that  $\mu \neq \delta_0$ . Let us define the function

$$M_\mu(z) = \int_{t \in \mathbb{R}} \frac{tz}{1 - tz} d\mu(t).$$

Then the  $R$ - and  $S$ -transforms<sup>6</sup> of  $\mu$ , denoted respectively by  $R_\mu$  and  $S_\mu$  are the analytic functions defined as follows

$$(2) \quad R_\mu(z) = [(1 + z)M_\mu^{-1}(z)]^{-1}, \quad S_\mu(z) = \frac{1 + z}{z} M_\mu^{-1}(z),$$

where the exponent  $^{-1}$  refers to the inversion of functions with respect to the operation of composition  $\circ$ . Note that  $M_\mu$  is an analytic function defined in  $\{z \in \mathbb{C}; 1/z \notin \text{support}(\mu)\}$ . Hence in the case where  $\mu$  is compactly supported, the functions  $M_\mu$  and  $(1 + z)M_\mu^{-1}(z)$  can be inverted in a neighborhood of zero as analytic functions in a neighborhood of zero vanishing at zero, with

---

<sup>6</sup>See the footnote 5.

non null derivative at zero. In the case where  $\mu$  is not compactly supported, these functions are inverted as functions on intervals  $(-\epsilon, 0)$  which are equivalent to  $(\text{positive constant}) \times z$  at zero [BV93].

Note that putting together both equations of (2), one gets

$$(3) \quad S_\mu(z) = \frac{1}{z} R_\mu^{-1}(z) = \frac{1+z}{z} M_\mu^{-1}(z).$$

The main properties of the  $R$ - and  $S$ -transforms are the fact that they characterize measures and their weak convergence and that they allow to compute free convolutions : for all  $\mu, \nu$ ,

$$(4) \quad R_{\mu \boxplus \nu} = R_\mu + R_\nu \quad \text{and} \quad S_{\mu \boxtimes \nu} = S_\mu S_\nu.$$

1.1.2. *The rectangular case: the rectangular  $R$ -transform with ratio  $\lambda$ .* In the same way, for  $\lambda \in [0, 1]$ , the rectangular free convolution with ratio  $\lambda$  can be computed with an analytic transform of probability measures. Let  $\nu$  be a symmetric probability measure on the real line. Let us define  $H_\nu(z) = z(\lambda M_{\nu^2}(z) + 1)(M_{\nu^2}(z) + 1)$ , where  $\nu^2$  denotes the push forward of  $\nu$  by the map  $t \mapsto t^2$ . Then with the same conventions about inverses of functions than in the previous section, the *rectangular  $R$ -transform with ratio  $\lambda$*  of  $\nu$  is defined to be

$$C_\nu(z) = U\left(\frac{z}{H_\nu^{-1}(z)} - 1\right),$$

where  $U(z) = \frac{-\lambda - 1 + [(\lambda + 1)^2 + 4\lambda z]^{1/2}}{2\lambda}$  for  $\lambda > 0$  and  $U(z) = z$  for  $\lambda = 0$ . By Theorems 3.8, 3.11 and 3.12 of [B09], the rectangular  $R$ -transform characterizes measures and their weak convergence, and for all pair  $\nu_1, \nu_2$  of symmetric probability measures,  $\nu_1 \boxplus_\lambda \nu_2$  is characterized by the fact that

$$(5) \quad C_{\nu_1 \boxplus_\lambda \nu_2} = C_{\nu_1} + C_{\nu_2}.$$

1.2. **A relation between the square and the rectangular  $R$ -transforms.** Let us fix  $\lambda \in [0, 1]$ . We recall that for any probability measure  $\rho$  on  $[0, +\infty)$ ,  $\sqrt{\rho}$  denotes the symmetrization of the push-forward of  $\rho$  by the map  $t \mapsto \sqrt{t}$  and that for  $\lambda > 0$ , we have defined  $\mu_\lambda$  to be the law of  $\lambda$  times a random variable with law the Marchenko-Pastur law with mean  $1/\lambda$ , i.e.  $\mu_\lambda$  is the law with support  $[(1 - \sqrt{\lambda})^2, (1 + \sqrt{\lambda})^2]$  and density

$$x \mapsto \frac{\sqrt{4\lambda - (x - 1 - \lambda)^2}}{2\pi\lambda x}.$$

For  $\lambda = 0$ , we let  $\mu_\lambda$  denote the Dirac mass at 1.

**Theorem 1.** *Let  $\mu$  be a probability measure on  $[0, +\infty)$ . Then we have*

$$R_\mu(z) = C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z).$$

**Remark 2** (The cumulants point of view). Suppose  $\mu$  to be compactly supported. Let us denote the free cumulants [NS06] of  $\mu$  by  $(k_n(\mu))_{n \geq 1}$  and the rectangular free cumulants with ratio  $\lambda$  [B09, B07b] of  $\sqrt{\mu \boxtimes \mu_\lambda}$  by  $(c_{2n}(\sqrt{\mu \boxtimes \mu_\lambda}))_{n \geq 1}$ . Then the previous theorem means that for all  $n \geq 1$ , one has

$$(6) \quad k_n(\mu) = c_{2n}(\sqrt{\mu \boxtimes \mu_\lambda}).$$

**Proof.** - First of all, note that by continuity of the applications  $\mu \mapsto \mu \boxtimes \mu_\lambda$ ,  $\rho \mapsto R_\rho$  and  $\nu \mapsto C_\nu$  with respect to weak convergence [BV93, B09], it suffices to prove the result in the case where  $\mu$  is compactly supported. In this case, the functions  $M_\mu, R_\mu, S_\mu, M_{\mu \boxtimes \mu_\lambda}, H_{\sqrt{\mu \boxtimes \mu_\lambda}}, C_{\sqrt{\mu \boxtimes \mu_\lambda}}$  are analytic in a neighborhood of zero and the operations of inversion on these functions or related ones can be used without precaution.

- If  $\lambda > 0$ , the free cumulants of the Marchenko-Pastur law with mean  $1/\lambda$  are all equal to  $1/\lambda$ , thus the ones of  $\mu_\lambda$  are given by the formula  $k_n(\mu_\lambda) = \lambda^{n-1}$  for all  $n \geq 1$  and  $R_{\mu_\lambda}(z) = \sum_{n \geq 1} \lambda^{n-1} z^n$ . From (3), it follows that  $S_{\mu_\lambda}(z) = \frac{1}{1+\lambda z}$ . Hence by (4), we have  $S_{\mu \boxtimes \mu_\lambda}(z) = \frac{S_\mu(z)}{1+\lambda z}$ , and by (3),

$$(7) \quad M_{\mu \boxtimes \mu_\lambda}(z) = \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1}.$$

Note that since  $\mu_0 = \delta_1$ , (7) stays true if  $\lambda = 0$ . Now, let us define the function  $T(x) = (\lambda x + 1)(x + 1)$ . Note that  $T(U(x - 1)) = x$  for  $x$  in a neighborhood of zero. We have

$$H_{\sqrt{\mu \boxtimes \mu_\lambda}}(z) = z \times T \circ M_{\mu \boxtimes \mu_\lambda}(z) = z \times T \circ \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1},$$

and

$$(8) \quad C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z) = U \left( \frac{z}{\left( z \times T \circ \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right)^{-1}} - 1 \right).$$

- Hence by (2) and (8), we have the following equivalence

$$\begin{aligned} R_\mu = C_{\sqrt{\mu \boxtimes \mu_\lambda}} &\iff ((z + 1)M_\mu^{-1}(z))^{-1} = U \left( \frac{z}{\left( z \times T \circ \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right)^{-1}} - 1 \right) \\ &\iff T \circ ((z + 1)M_\mu^{-1}(z))^{-1} = \frac{z}{\left( z \times T \circ \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right)^{-1}} \\ &\iff \left( z \times T \circ \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right)^{-1} \times T \circ ((z + 1)M_\mu^{-1}(z))^{-1} = z. \end{aligned}$$

Composing both terms on the right by  $(z + 1)M_\mu^{-1}(z)$ , it gives

$$R_\mu = C_{\sqrt{\mu \boxtimes \mu_\lambda}} \iff \left( z \times T \circ \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right)^{-1} \circ ((z + 1)M_\mu^{-1}(z)) \times T(z) = (z + 1)M_\mu^{-1}(z).$$

Dividing by  $T(z)$ , it gives

$$\begin{aligned}
R_\mu = C_{\sqrt{\mu \boxtimes \mu_\lambda}} &\iff \left( z \times T \circ \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right)^{-1} \circ ((z + 1)M_\mu^{-1}(z)) = \frac{M_\mu^{-1}(z)}{1 + \lambda z} \\
&\iff (z + 1)M_\mu^{-1}(z) = \left( z \times T \circ \left( \frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right) \circ \frac{M_\mu^{-1}(z)}{1 + \lambda z} \\
&\iff (z + 1)M_\mu^{-1}(z) = \frac{M_\mu^{-1}(z)}{1 + \lambda z} T(z),
\end{aligned}$$

which is obviously true by definition of  $T(z)$ .  $\square$

**1.3. Main result of the paper.** The main theorem of this paper is the following one.  $\lambda \in [0, 1]$  is still fixed.

**Theorem 3.** *For any pair  $\mu, \mu'$  of probability measures on  $[0, +\infty)$ , we have*

$$(9) \quad \sqrt{\mu \boxtimes \mu_\lambda} \boxplus_\lambda \sqrt{\mu' \boxtimes \mu_\lambda} = \sqrt{(\mu \boxplus \mu') \boxtimes \mu_\lambda}.$$

**Remark 4.** Note that in the case where  $\lambda = 0$ , this theorem expresses what we already knew about  $\boxplus_0$  (and which is explained in the second paragraph of the introduction), but that the case  $\lambda = 1$  isn't a consequence of the already known formula  $\boxplus_1 = \boxplus$ .

**Remark 5.** Part of this theorem could have been deduced from Theorem 6 of [DR07]. However, (9) could be deduced from the theorem of Debbah and Ryan only for laws  $\mu, \mu'$  which can be expressed as limit singular laws of  $n$  by  $p$  (for  $n/p \simeq \lambda$ ) corners of large  $p \times p$  bi-unitarily invariant random matrices, but it follows from Theorem 14.10 of [NS06] that not every law has this form. Moreover, even though the idea which led us to our result was picked in the pioneer work of Debbah and Ryan, our proof is much shorter and shows the connection with the rectangular machinery in a more clear way (via Theorem 1 and Remark 2).

**Proof.** Define  $\nu := \sqrt{\mu \boxtimes \mu_\lambda} \boxplus_\lambda \sqrt{\mu' \boxtimes \mu_\lambda}$ . By (5), we have

$$C_\nu = C_{\sqrt{\mu \boxtimes \mu_\lambda}} + C_{\sqrt{\mu' \boxtimes \mu_\lambda}}.$$

Thus, by Theorem 1, and (4), we have

$$C_\nu = R_\mu + R_{\mu'} = R_{\mu \boxplus \mu'} = C_{\sqrt{(\mu \boxplus \mu') \boxtimes \mu_\lambda}}.$$

Hence by injectivity of the rectangular  $R$ -transform (Theorem 3.8 of [B09]), (9) is valid.  $\square$

The formula (9) gives us a new insight on rectangular free convolutions: it allows to express it, in certain cases, in terms of the free convolutions “of square type”  $\boxplus$  and  $\boxtimes$ . However, only laws which can be expressed under the form

$$(10) \quad \sqrt{\mu \boxtimes \mu_\lambda}, \quad (\mu \text{ probability measure on } [0, +\infty))$$

can have their rectangular convolution computed via formula (9). Thus it seems natural to ask whether all symmetric laws can be expressed like in (10). Note that it is equivalent to the fact that any law on  $[0, +\infty)$  can be expressed under the form  $\mu \boxtimes \mu_\lambda$ , which is equivalent to the fact that the Dirac mass at one  $\delta_1$  can be expressed under the form  $\mu \boxtimes \mu_\lambda$ . Indeed, if  $\delta_1 = \mu \boxtimes \mu_\lambda$ , then any law  $\tau$  on  $[0, +\infty)$  satisfies  $\tau = \tau \boxtimes \delta_1 = (\tau \boxtimes \mu) \boxtimes \mu_\lambda$ . The following proposition shows that it is not the case. However, Theorem 11 will show that many symmetric laws can be expressed like in (10).

**Proposition 6.** *Unless  $\lambda = 0$ , the law  $\frac{\delta_1 + \delta_{-1}}{2}$  cannot be expressed under the form  $\sqrt{\mu \boxtimes \mu_\lambda}$  for  $\mu$  probability measure on  $[0, +\infty)$ .*

**Proof.** Suppose that  $\lambda > 0$  and that there is a probability measure  $\mu$  on  $[0, +\infty)$  such that  $\frac{\delta_1 + \delta_{-1}}{2} = \sqrt{\mu \boxtimes \mu_\lambda}$ . Then  $\delta_1 = \mu \boxtimes \mu_\lambda$ . This is impossible, by Corollary 3.4 of [B06], which states that the free multiplicative convolution of two laws which are not Dirac masses has always a non null absolutely continuous part (there is another, more direct way to see that it is impossible: by (4), such a law  $\mu$  has to satisfy  $S_\mu(z) = 1 + \lambda z$ , which implies that for  $z$  small enough,  $M_\mu(z) = \frac{z - 1 + [(1-z)^2 + 4\lambda z]^{1/2}}{2\lambda}$ : such a function doesn't admit any analytic continuation to  $\mathbb{C} \setminus [0, +\infty)$ , thus no such probability measure  $\mu$  exists).  $\square$

Theorem 3 has a consequence on the free convolutions “of square type” which wasn't known yet, despite the many papers written the last years about questions related to the arithmetics of these convolutions, e.g. [BV95, BPB99, B04, CG08a, CG08b, BBG08, BBCC08].

**Corollary 7.** *For any pair  $\mu, \mu'$ , of probability measures on  $[0, +\infty)$  we have*

$$(11) \quad \sqrt{\mu \boxtimes \mu_1} \boxplus \sqrt{\mu' \boxtimes \mu_1} = \sqrt{(\mu \boxplus \mu') \boxtimes \mu_1}.$$

**Proof.** It is an obvious consequence of Theorem 3 and of the fact that  $\boxplus_1 = \boxplus$ .  $\square$

**Remark 8.** The referee of the paper communicated to us a proof of (11) which is not, as ours, based on computations on the  $R$ - and  $S$ -transforms, but on the direct proof of (6) in the special case  $\lambda = 1$ . Let us briefly outline this proof. When  $\lambda = 1$ , by [B07a, Eq. (4.1)], (6) reduces to

$$(12) \quad k_n(\mu) = k_{2n}(\sqrt{\mu \boxtimes \mu_1}).$$

Let  $a, s$  are free elements in a tracial non commutative probability space with respective distributions  $\mu$  and the standard semicircle law. By [NS06, Prop. 12.13],  $s^2$  has distribution  $\mu_1$ , hence  $sas$  has distribution  $\mu \boxtimes \mu_1$ . It follows, by [NS06, Prop. 12.18], that for all  $n$ , the  $n$ -th moment of  $\mu$  is equal to  $k_n(\mu \boxtimes \mu_1)$ . But by [NS06, Prop. 11.25], for all  $n$ , we have

$$k_n(\mu \boxtimes \mu_1) = \sum_{\pi \in \text{NC}(n)} \prod_{V \in \pi} k_{2|V|}(\sqrt{\mu \boxtimes \mu_1}).$$

It follows, using the expression of the  $n$ -th moment of  $\mu$  in terms of its free cumulants, that for all  $n$ ,

$$\sum_{\pi \in \text{NC}(n)} \prod_{V \in \pi} k_{|V|}(\mu) = \sum_{\pi \in \text{NC}(n)} \prod_{V \in \pi} k_{2|V|}(\sqrt{\mu \boxtimes \mu_1}),$$

and that for all  $n$ ,  $k_n(\mu) = k_{2n}(\sqrt{\mu \boxtimes \mu_1})$ .

## 2. CONSEQUENCES ON SQUARE AND RECTANGULAR INFINITE DIVISIBILITY

**2.1. Prerequisites on infinite divisibility and Lévy-Kinchine formulas.** Infinite divisibility is a fundamental probabilistic notion, at the base of Lévy processes, and which allows to explain deep relations between limit theorems for sums of either independent random variables, square or rectangular random matrices. Let us briefly recall basics of this theory [GK54, Sa99, BV93, BPB99, B07a].

Let  $*$  denote the classical convolution of probability measures on the real line. Firstly, recall that a probability measure  $\mu$  is said to be  $*$ -infinitely divisible (resp.  $\boxplus$ -,  $\boxplus_\lambda$ -infinitely divisible) if for all integer  $n$ , there exists a probability measure  $\nu_n$  such that  $\nu_n^{*n} = \mu$  (resp.  $\nu_n^{\boxplus n} = \mu$ ,  $\nu_n^{\boxplus_\lambda n} = \mu$ ). In this case, there exists a  $*$ - (resp.  $\boxplus$ -,  $\boxplus_\lambda$ -) semigroup  $(\mu_t)_{t \geq 0}$  such that  $\mu_0 = \delta_0$

and  $\mu_1 = \mu$ . For all  $t$ ,  $\mu_t$  is denoted by  $\mu^{*t}$  (resp.  $\mu^{\boxplus t}, \mu^{\boxplus_\lambda t}$ ). Infinitely divisible distributions have been classified:  $\mu$  is  $*$ - (resp.  $\boxplus$ -) infinitely divisible if and only if there exists a real number  $\gamma$  and a positive finite measure on the real line  $\sigma$  such that the Fourier transform is  $\hat{\mu}(t) = \exp \left[ i\gamma t + \int_{\mathbb{R}} (e^{itx} - 1 - \frac{itx}{x^2+1}) \frac{x^2+1}{x^2} d\sigma(x) \right]$  (resp.  $R_\mu(z) = \gamma z + z \int_{\mathbb{R}} \frac{z+t}{1-tz} d\sigma(t)$ ). Moreover, in this case, such a pair  $(\gamma, \sigma)$  is unique, it is called the *Lévy pair* of  $\mu$  and we denote  $\mu$  by  $\nu_*^{\gamma, \sigma}$  (resp.  $\nu_{\boxplus}^{\gamma, \sigma}$ ). For all  $t \geq 0$ ,  $\mu_t$  has Lévy pair  $(t\gamma, t\sigma)$ . In the same way, a symmetric probability measure  $\nu$  is  $\boxplus_\lambda$ -infinitely divisible if and only if there exists a positive finite symmetric measure on the real line  $G$  such that  $C_\nu(z) = z \int_{\mathbb{R}} \frac{1+t^2}{1-zt^2} dG(t)$ . In this case, the measure  $G$  is unique, and  $\nu$  will be denoted by  $\nu_{\boxplus_\lambda}^G$ . The correspondences  $\nu_*^{\gamma, \sigma} \longleftrightarrow \nu_{\boxplus}^{\gamma, \sigma}$  (for any pair  $(\gamma, \sigma)$  as above) and  $\nu_*^{0, G} \longleftrightarrow \nu_{\boxplus_\lambda}^G$  (for any  $G$  as above) are called *Bercovici-Pata bijections*. These bijections have many deep properties [BPB99, B07a], some of which will be mentioned in the proof of the following lemma.

**Lemma 9.** *Let  $\gamma$  be a real number and  $\sigma$  be a positive finite measure on the real line. Then we have equivalence between:*

- (i) *For all  $t \geq 0$ ,  $\nu_*^{t\gamma, t\sigma}$  is supported on  $[0, +\infty)$ .*
- (ii) *For all  $t \geq 0$ ,  $\nu_{\boxplus}^{t\gamma, t\sigma}$  is supported on  $[0, +\infty)$ .*
- (iii) *We have  $\sigma((-\infty, 0]) = 0$  and the integral  $\int \frac{1}{x} d\sigma(x)$  is finite and  $\leq \gamma$ .*

**Proof.** The equivalence between (i) and (iii) follows from Theorem 24.7 and Corollary 24.8 of [Sa99]. Let us prove the equivalence between (i) and (ii). In order to do that, let us recall a fact proved in [BPB99]: for any Lévy pair  $(\gamma, \sigma)$  and any sequence  $(\nu_n)$  of probability measures, one has

$$(13) \quad \nu_n^{*n} \text{ converges weakly to } \nu_*^{\gamma, \sigma} \iff \nu_n^{\boxplus n} \text{ converges weakly to } \nu_{\boxplus}^{\gamma, \sigma}.$$

Let us suppose (i) (resp. (ii)) to be true. Let us fix  $t \geq 0$ . For all  $n$ , we have

$$\left(\nu_*^{\frac{t\gamma}{n}, \frac{t\sigma}{n}}\right)^{*n} = \nu_*^{t\gamma, t\sigma} \quad (\text{resp. } \left(\nu_{\boxplus}^{\frac{t\gamma}{n}, \frac{t\sigma}{n}}\right)^{\boxplus n} = \nu_{\boxplus}^{t\gamma, t\sigma}).$$

Thus by (13),

$$\left(\nu_*^{\frac{t\gamma}{n}, \frac{t\sigma}{n}}\right)^{\boxplus n} \text{ converges weakly to } \nu_{\boxplus}^{t\gamma, t\sigma} \quad (\text{resp. } \left(\nu_{\boxplus}^{\frac{t\gamma}{n}, \frac{t\sigma}{n}}\right)^{*n} \text{ converges weakly to } \nu_*^{t\gamma, t\sigma}).$$

Thus since any free (resp. classical) additive convolution and any weak limit of measures with supports on  $[0, +\infty)$  has support on  $[0, +\infty)$ , (ii) (resp. (i)) holds.  $\square$

**Remark 10.** Note that (i) is equivalent to the fact that there exists  $t > 0$  such that  $\nu_*^{t\gamma, t\sigma}$  is supported on  $[0, +\infty)$  [Sa99, Cor. 24.8]. However, the same is not true for the free infinitely divisible laws. Indeed, let, for each  $t \geq 0$ ,  $\text{MP}_t$  denote the Marchenko-Pastur law with mean  $t$  [HP00, Ex. 3.3.5] and let us define, for each  $t$ ,  $\mu_t = \text{MP}_t * \delta_{-t/4}$ . Then since free and classical convolutions with Dirac masses are the same,  $(\mu_t)_{t \geq 0}$  is a convolution semi-group with respect to  $\boxplus$ . But  $\mu_4$  is supported on  $[0, +\infty)$ , whereas for each  $t \in (0, 1]$ , the support of  $\mu_t$  contains a negative number (namely  $-t/4$ ).

**2.2. Main result of the section.** The following theorem allows us to claim that even though not every symmetric law can be expressed under the form  $\sqrt{\mu \boxtimes \mu_\lambda}$  for  $\mu$  law on  $[0, +\infty)$  (see Proposition 6), many of them have this form.  $\lambda \in [0, 1]$  is still fixed.

For  $G$  measure on the real line, we let  $G^2$  denote the push-forward of  $G$  by the function  $t \mapsto t^2$ .



**Theorem 11.** (i) Let  $\mu$  be a  $\boxplus$ -infinitely divisible law such that for all  $t \geq 0$ ,  $\mu^{\boxplus t}$  is supported on  $[0, +\infty)$ . Then the law  $\sqrt{\mu \boxtimes \mu_\lambda}$  is  $\boxplus_\lambda$ -infinitely divisible, with Lévy measure the only symmetric measure  $G$  such that

$$(14) \quad G^2 = \left( \gamma - \int \frac{1}{x} d\sigma(x) \right) \delta_0 + \frac{1+x^2}{x(1+x)} d\sigma(x),$$

where  $(\gamma, \sigma)$  denotes the Lévy pair of  $\mu$ .

(ii) Reciprocally, any  $\boxplus_\lambda$ -infinitely divisible law  $\nu$  has the form  $\sqrt{\mu \boxtimes \mu_\lambda}$  for some  $\boxplus$ -infinitely divisible law  $\mu$  such that for all  $t \geq 0$ ,  $\mu^{\boxplus t}$  is supported on  $[0, +\infty)$ . Moreover, the Lévy pair  $(\gamma, \sigma)$  of  $\mu$  is defined by

$$(15) \quad \gamma = \int_{[0, +\infty)} \frac{1+x}{1+x^2} dG^2(x) \quad \text{and} \quad \sigma = \frac{x(1+x)}{1+x^2} dG^2(x),$$

where  $G$  denotes the Lévy measure of  $\nu$ .

**Proof.** (i) Note that by Theorem 3, the map  $\mu \mapsto \sqrt{\mu \boxtimes \mu_\lambda}$  is a morphism from the set of laws on  $[0, +\infty)$  to the set on symmetric laws on the real line endowed respectively with the operations  $\boxplus$  and  $\boxplus_\lambda$ . Thus if  $\mu$  is  $\boxplus$ -infinitely divisible, then  $\sqrt{\mu \boxtimes \mu_\lambda}$  is  $\boxplus_\lambda$ -infinitely divisible. Moreover, if the Lévy pair of  $\mu$  is  $(\gamma, \sigma)$ , then its  $R$ -transform is  $R_\mu(z) = \gamma z + z \int_{t \in \mathbb{R}} \frac{z+t}{1-zt} d\sigma(t)$ . By Theorem 1, it implies that  $C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z) = \gamma z + z \int_{t \in \mathbb{R}} \frac{z+t}{1-zt} d\sigma(t)$ . But by uniqueness, the Lévy measure  $G$  of  $\sqrt{\mu \boxtimes \mu_\lambda}$  is characterized by the fact that  $C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z) = z \int_{\mathbb{R}} \frac{1+t^2}{1-zt^2} dG(t)$ . Thus to prove (14), it suffices to prove that for  $G$  given by (14), for all  $z$ , one has

$$\gamma z + z \int_{t \in \mathbb{R}} \frac{z+t}{1-zt} d\sigma(t) = z \int_{\mathbb{R}} \frac{1+t^2}{1-zt^2} dG(t),$$

which can easily be verified.

(ii) Let  $\nu$  be a  $\boxplus_\lambda$ -infinitely divisible law with Lévy measure denoted by  $G$ . Let  $(\gamma, \sigma)$  be the Lévy pair defined by (15). Note that  $(\gamma, \sigma)$  satisfies (iii) of Lemma 9, thus, for  $\mu := \nu_{\boxplus}^{\gamma, \sigma}$ , for all  $t \geq 0$ , the law  $\mu^{\boxplus t}$  is actually supported by  $[0, +\infty)$ . Thus by (i),  $\sqrt{\mu \boxtimes \mu_\lambda}$  is  $\boxplus_\lambda$ -infinitely divisible with Lévy measure the only symmetric measure  $H$  satisfying

$$H^2 = \left( \gamma - \int \frac{1}{x} d\sigma(x) \right) \delta_0 + \frac{1+x^2}{x(1+x)} d\sigma(x).$$

To prove that  $\sqrt{\mu \boxtimes \mu_\lambda} = \nu$ , it suffices to prove that  $H = G$ , which can easily be verified.  $\square$

One of the consequences of this theorem is that it gives us a description of the free multiplicative convolution of two Marchenko-Pastur laws (i.e. free Poisson laws), one of them having a mean  $\geq 1$ . For all  $t > 0$ , the Marchenko-Pastur law  $\text{MP}_t$  with mean  $t$  has been introduced at Remark 10.

**Corollary 12.** Consider  $a, c > 0$  such that  $a > 1$ . Then  $\text{MP}_c \boxtimes \text{MP}_a$  is the push forward, by the map  $x \mapsto ax^2$ , of the  $\boxplus_\lambda$ -infinitely divisible law with Lévy measure  $\frac{c}{4}(\delta_1 + \delta_{-1})$  for  $\lambda = 1/a$ .

**Proof.** It suffices to notice that for  $\lambda = 1/a$ ,  $\text{MP}_a$  is the push-forward, by the map  $x \mapsto ax$ , of the law  $\mu_\lambda$ , that  $\text{MP}_c$  is the  $\boxplus$ -infinitely divisible law with Lévy pair  $(c/2, c/2\delta_1)$ , and then to apply (i) of Theorem 11.  $\square$

This corollary can be interpreted as the coincidence of the limit laws of two different matrix models. Indeed, the  $\boxplus_\lambda$ -infinitely divisible law with Lévy measure  $\frac{c}{4}(\delta_1 + \delta_{-1})$  was already

known [B07a, Prop. 6.1] to be the limit symmetrized singular law of the random matrix  $M := \sum_{k=1}^p u_k v_k^*$ , for  $n, p, q$  tending to infinity in such a way that  $p/n \rightarrow c$  and  $n/q \rightarrow \lambda$  and  $(u_k)_{k \geq 1}, (v_k)_{k \geq 1}$  two independent families of independent random vectors such that for all  $k$ ,  $u_k, v_k$  are uniformly distributed on the unit spheres of respectively  $\mathbb{C}^n, \mathbb{C}^q$ . Thus, if, for large  $n, p, q$ 's such that  $p/n \simeq c$  and  $n/q \simeq \lambda$ , one considers such a random matrix  $M$  and also two independent random matrices  $T, Q$  with respective dimensions  $n \times p, n \times q$ , the entries of which are independent real standard Gaussian random variables, then the empirical spectral measures of the random matrices  $MM^*$  and  $\frac{1}{nq}TT^*QQ^*$  are close to each other, as illustrated by Figure 1.

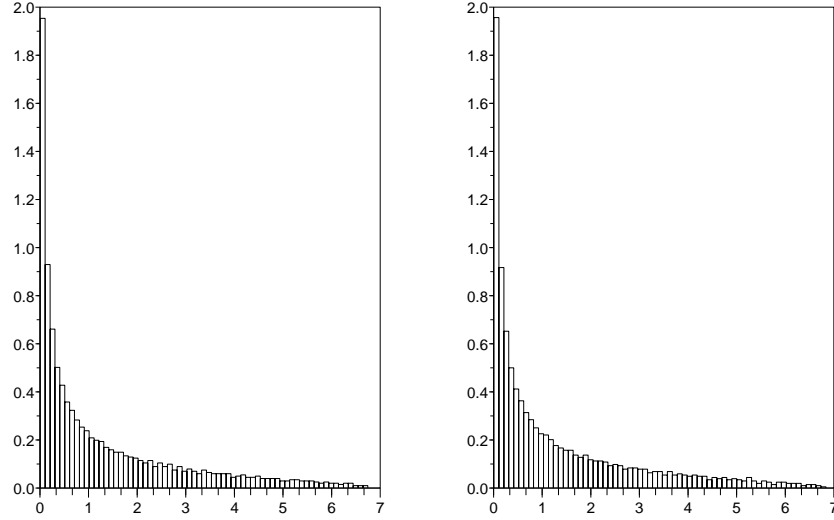


FIGURE 1. Histograms of the spectrums of  $MM^*$  (left) and  $\frac{1}{nq}TT^*QQ^*$  (right) for  $n = 2000$ ,  $\lambda = 0.6$ ,  $c = 1.3$ .

## REFERENCES

- [BBCC08] Banica, T., Belinschi, S., Capitaine, M., Collins, B. *Free Bessel Laws*, to appear in Canad. J. of Math.
- [B06] Belinschi, S. *A note on regularity for free convolutions*. Ann. Inst. H. Poincaré Probab. Statist. 42 (2006), no. 5, 635–648.
- [BBG08] Belinschi, S., Benaych-Georges, F., Guionnet, A. *Regularization by free additive convolution, square and rectangular cases*. 2008, to appear in Complex Analysis and Operator Theory.
- [B04] Benaych-Georges, F. *Failure of the Raikov theorem for free random variables*. Séminaire de Probabilités XXXVIII, p. 313–320 (2004)
- [B07a] Benaych-Georges, F. *Infinitely divisible distributions for rectangular free convolution: classification and matricial interpretation* Probability Theory and Related Fields. Volume 139, Numbers 1–2 / september 2007, 143–189.
- [B07b] Benaych-Georges, F. *Rectangular random matrices, related free entropy and free Fisher’s information*. 2007, to appear in Journal of Operator Theory.
- [B09] Benaych-Georges, F. *Rectangular random matrices, related convolution*. Probability Theory and Related Fields. Volume 144, Numbers 3–4 / july 2009, 471–515.
- [BPB99] Bercovici, H., Pata, V., with an appendix by Biane, P. *Stable laws and domains of attraction in free probability theory* Annals of Mathematics, 149. (1999) 1023–1060
- [BV93] Bercovici, H., Voiculescu, D. *Free convolution of measures with unbounded supports* Indiana Univ. Math. J. 42 (1993) 733–773

- [BV95] Bercovici, H., Voiculescu, D. *Superconvergence to the central limit and failure of the Cramér theorem for free random variables*. Probability Theory and Related Fields 102 (1995) 215–222
- [CG08a] Chistyakov, G. P., Götze, F. *Limit theorems in free probability theory. I*. Ann. Probab. 36 (2008), no. 1, 54–90.
- [CG08b] Chistyakov, G. P., Götze, F. *Limit theorems in free probability theory. II*. Cent. Eur. J. Math. 6 (2008), no. 1, 87–117.
- [DR07] Debbah, M., Ryan, Ø. *Multiplicative free Convolution and Information-Plus-Noise Type Matrices*. arXiv. The submitted version of this paper, more focused on applications than on the result we are interested in here, is [DR08].
- [DR08] Debbah, M., Ryan, Ø. *Free Deconvolution for Signal Processing Applications* Submitted.
- [GK54] Gnedenko, V., Kolmogorov, A.N. *Limit distributions for sums of independent random variables* Addison-Wesley Publ. Co., Cambridge, Mass., 1954
- [HP00] Hiai, F., Petz, D. *The semicircle law, free random variables, and entropy* Amer. Math. Soc., Mathematical Surveys and Monographs Volume 77, 2000
- [NS06] Nica, A., Speicher, R. *Lectures on the combinatorics of free probability*. London Mathematical Society Lecture Note Series, 335. Cambridge University Press, Cambridge, 2006.
- [Sa99] Sato, K.I. *Lévy processes and infinitely divisible distributions* Volume 68 of Cambridge Studies in Advanced Mathematics, Cambridge University Press, Cambridge, 1999
- [VDN91] Voiculescu, D.V., Dykema, K., Nica, A. *Free random variables* CRM Monographs Series No.1, Amer. Math. Soc., Providence, RI, 1992

FLORENT BENAYCH-GEORGES, LPMA, UPMC UNIV PARIS 6, CASE COURIER 188, 4, PLACE JUSSIEU, 75252 PARIS CEDEX 05, FRANCE, AND CMAP, ÉCOLE POLYTECHNIQUE, ROUTE DE SACLAY, 91128 PALAISEAU CEDEX, FRANCE

*E-mail address:* florent.benaych@gmail.com